

## The deluge of data

Daily By 2020

verage internet user 1.5 GB

Autonomous vehicle 13

ONNECTED AIRPLANES TB

Smart Factory PB

Cloud video Provider 50 pB



Business Insights

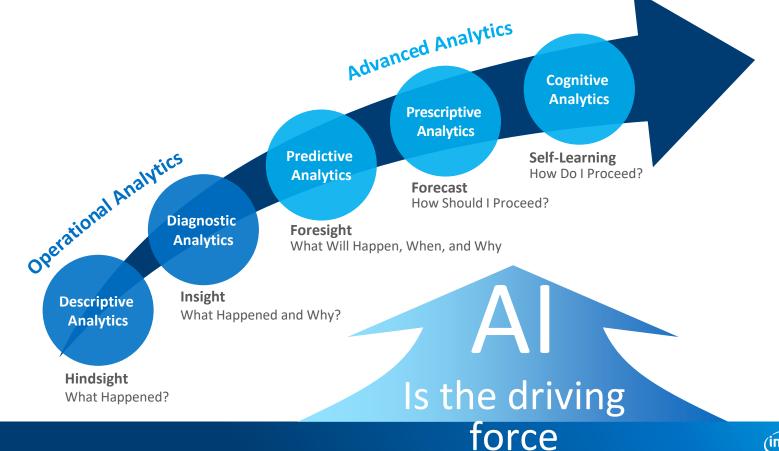
Operational Insights

Security Insights

Source: Amalgamation of analyst data and Intel analysis



## The path to deeper insight



## Al will transform















In-Vehicle

Experience

**Automated** 

Driving

Aerospace

Shipping

Search & Rescue





Consum er

**Smart Assistants Enhanced** Diagnostics Chatbots Drug Search Discovery

Personalization

Augmented Reality

Robots

**Health Finance** Retail

Algorithmic

Fraud Detection

Patient Care

Research

Sensory Aids

Trading

Research

Personal Finance

**Risk Mitigation** 

Support

Experience

Marketing

Merchandising

Loyalty

Supply Chain

Security

Defense

Govern

ment

Data Insights

Safety & Security

Resident Engagement

Smarter Cities

**Energy** 

Oil & Gas **Exploration** 

> Smart Grid

Operational Improvement

Conservation

Transpo Industri

al

Factory Automation

**Predictive** Maintenance

Precision Agriculture

**Field Automation** 

Advertising

Other

Education

Gaming

Professional & IT Services

Telco/Media

Sports

Source: Intel forecast



## Ai adoption is nascent

According to a recent Forrester Research survey... 58% of business and technology professionals said they're researching AI, but only said they are currently using AI systems.

## Al Opportunity assessment

Brainstorm and Prioritize Business Challenges

What business challenges am I facing today?

What business value is tied to each challenge?

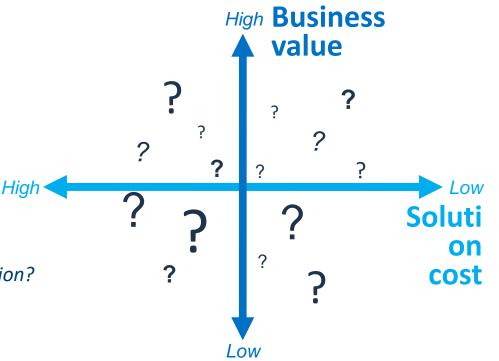
What are my solution requirements?

What data do I have at my disposal?

Do I know how to approach each challenge?

Do I have what I need to implement each solution?

How costly is it to implement each solution?



## Which approach is right?

A large **manufacturer** uses data to improve their operations, with each challenge using a different approach to deliver maximum business value at the lowest possible cost

| Challenge                               | Best approach   | Approac<br>h                           | Answer  |
|---|---|--|---|
| How many widgets should we manufacture? | Analyze historical supply/demand                              | Analytics/<br>Business<br>Intelligence | 10,000  |
| What will our yield be?                 | Algorithm that correlates many variables to yield             | Statistical/<br>Machine<br>Learning    | At current conditions, yield will be at 90% with 10% loss expected                        |
| Which widgets have visual defects?      | Algorithm that learns to identify defects in images           | Deep<br>Learning                       | Widget 1003, Widget 1094  |
| How do I resolve each defect?           | Algorithms that learns associations between defect type/cause | Reasoning<br>Systems                   | Widget 1003 – Tolerance issue, calibration Widget 1094 – Etch issue, gas pressure warning |

Learn More in the Next Secti on

## The AI lifecycle

#### **Define the Challenge**

#### **Culture & Resources**

Organization embraces data insights, sponsors properly resourced teams, and prioritizes analytic development work

#### Infrastructure

Organization secures hardware and software infrastructure that supports data processing in a timely manner

#### **Source Data**

HITIH

\*

Team understands and obtains the right data that explains the business problem to achieve results

#### **Problem Solving Process**

Team breaks down the defined business problem into workable steps to translate the right data to achieve results

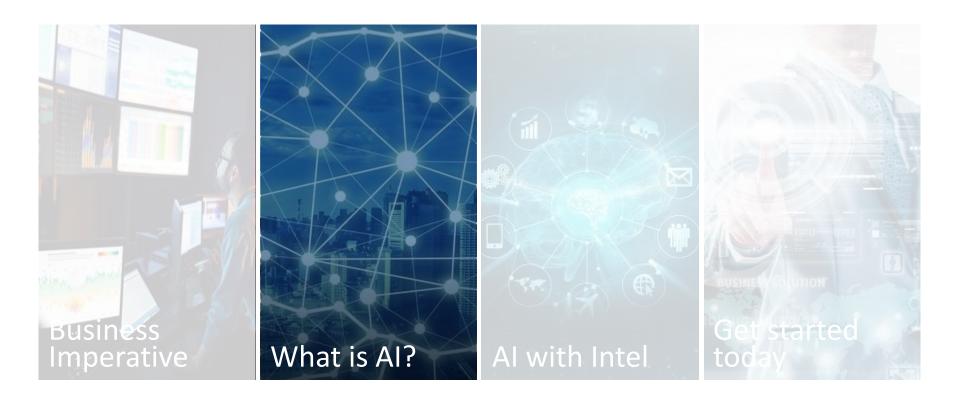
#### **Expertise**

A team of management sponsors, data scientists, data engineers, solution architects, and domain experts identifies the right data and works to translate the data to achieve results

### Philosophy

Team embraces fail-fast continuous improvement practices to evaluate their success in translating data to achieve results





## Artificial Intelligence

is the ability of machines to learn from experience, without explicit programming, in order to perform cognitive functions associated with the human mind

### Artificial Intelligence

### Machine learning

Algorithms whose performance improve as they are exposed to more data over time

## Deep learning

Subset of machine learning in which multi-layered neural networks learn from vast amounts of data

# ser look Deep Learning



## Machine learning



Algorithms designed to deliver better insight with more data

**Regression** (Linear/Logistic)

**Classification** (Support Vector Machines/SVM, Naïve Bayes)

Clustering (Hierarchical, Bayesian, K-Means, DBSCAN)

**Decision Trees** (RandomForest)

**Extrapolation** (Hidden Markov Models/HMM)

More...

Neural networks used to infer meaning from large dense datasets

Image Recognition (Convolutional Neural Networks/CNN. Single-Shot Detector/SSD)

**Speech Recognition** (Recurrent Neural Network/RNN)

Natural Language Processing (Long-Short Term Memory/LSTM)

**Data Generation** (Generative Adversarial Networks/GAN)

**Recommender System** (Multi-Layer Perceptron/MLP)

Time-Series Analysis (LSTM, RNN)

Reinforcement Learning (CNN, RNN)

More...

**Hybrid of analytics & AI** techniques designed to find meaning in diverse datasets

Associative Memory (Intel® Saffron Al memory base)

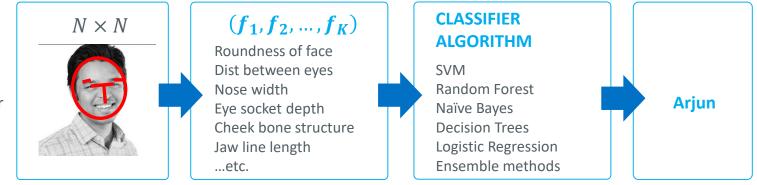
← See also: machine & deep learning techniques

More...

## Machine vs. Deep learning

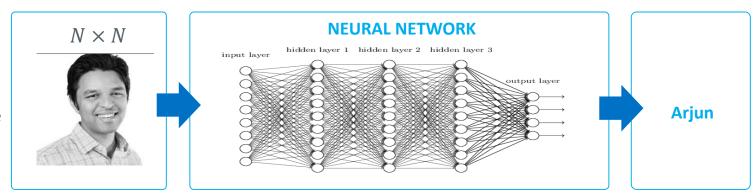
## Machine Learning

How do you engineer the best features?



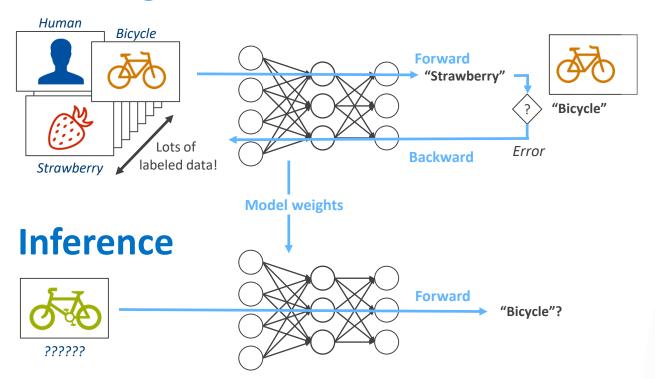
### Deep Learning

How do you guide the model to find the best features?



## Deep learning Basics

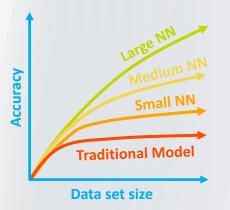
### **Training**





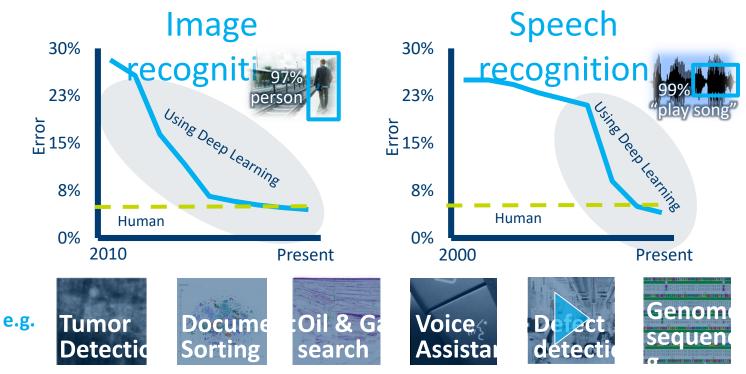
### Did you

Training with which was data set AND deep (many layered) neural network often leads to the highest accuracy inference



## Deep learning breakthroughs

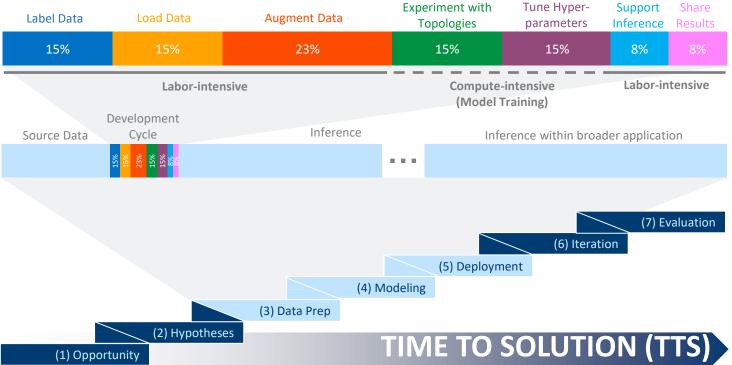
Machines able to meet or exceed human image & speech recognition



Source: ILSVRC ImageNet winning entry classification error rate each year 2010-2016 (Left), https://www.microsoft.com/en-us/research/blog/microsoft-researchers-achieve-new-conversational-speech-recognition-milestone/ (Right)

## The Journey to Production Al Defect

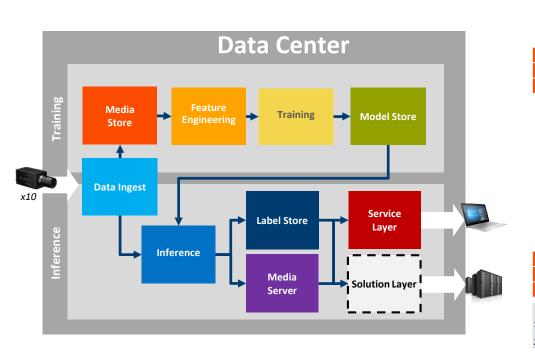


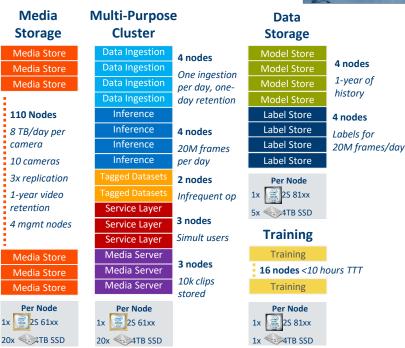


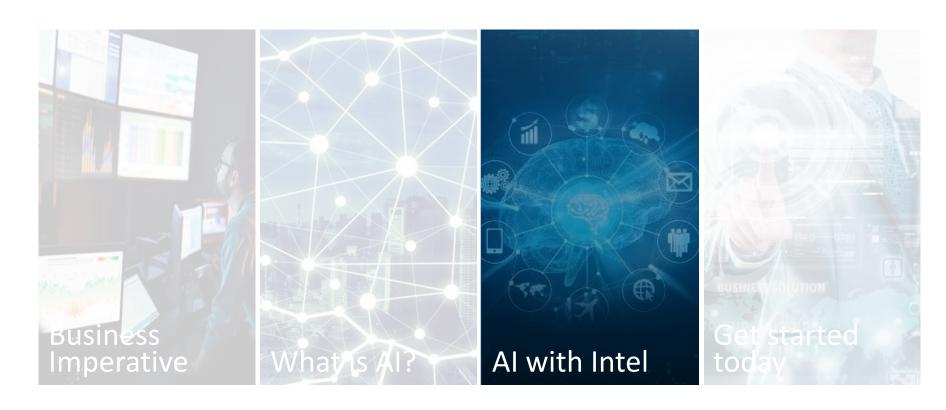
## Deep learning in practice

Al deployments have many interconnected parts











#### Smarter AI Through the Industry's Most Comprehensive Platform

### **Data**

Intel analytics ecosystem to get your data ready from integration to analysis



### **Future**

Driving AI forward through R&D, investment and policy leadership

## da

Intel analytics ecosystem to get your data ready from integration to analysis



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Source(s)?
Structured?
Volume?
JRABILITY?
Streaming?
LOCALITY?
ERNANCE?
Other?

Tool for live streaming data ingestion from Internet of Things (IOT) sensors in endpoint devices

e.g. Kafka\*, Sqoop\*, MQTT\*, WS\*, REST\*, Flume\* File, block or objectbased storage solution given cost, access, volume and performance requirements

e.g. Lustre\*, IBM\*
Spectrum\* Scale\* (GPFS),
Dell/EMC\* Isilon\*,
MySQL\* (OLTP),
Tera\*data\* (EDW), AWS\*
S3\* (ODS), HDFS\* (NoSQL), Hbase\* (In-Mem
DB)

Integration, cleaning, normalization and other transformations on batch and/or streaming data

e.g. Hadoop\* MapReduce\*, Apache\* Storm\*, Beam\* Job scheduling and storage management framework for distributed computation in various domains

e.g. SLURM\*, PBS\*, YARN\*, Mesos\*, Kubernetes\* Applications in HPC, Big Data, HPDA, AI & more that have access to a common compute and data pool

e.g. MPI\*, SHMEM\*, Hadoop\*, Spark\*, Apache\*, Flink\*, TensorFlow\*, MXNet\*

\*Other names and brands may be claimed as the property of others. Non-exhaustive list of offerings in each category.

## solution Sartner ecosystem to facilitate Al in nance, health, retail, industrial & more



## Intel Al Builders



Your one-stop-shop to find systems, software and solutions using Intel® AI technologies

builders.intel.com/ai/membership

## **Reference** solutions



Get a head start using the many case studies, solution briefs and more reference collaterals spanning multiple applications

builders.intel.com/ai/solutionslibrary

**SEE ALSO:** Al Solution Deck (internal)



### Portfolio of software tools to accelerate time-to-solution



### **TOOLKITS**

**Developers** 



#### **DEEP LEARNING DEPLOYMENT**

#### **OpenVINO™**

Open Visual Inference & Neural Network Optimization toolkit for inference deployment on CPU/GPU/FPGA for TF, Caffe\* & MXNet\*

#### Intel<sup>®</sup> Movidius™ SDK

Optimized inference deployment on Intel VPUs for TensorFlow\* & Caffe\*

#### **REASONING**

#### Intel® Saffron™ AI

Cognitive solutions on CPU for anti-money laundering, predictive maintenance, more



Open-source tool to compress deep learning development cycle

### libraries

Data Scientists



#### **MACHINE LEARNING LIBS**

Random

#### **Pvthon**

- Scikit-learn
- Pandas
- NumPy
- Cart
- Distributed MILib (on Spark)
- Mahout



**Now optimized for CPU** 





**DEEP LEARNING FRAMEWORKS** 

Caffe2 PYTORCH CNTK



**Forest** • e1071

TensorFlow\* MXNet\*

Caffe\* BigDL/Spark\*

**cIDNN** 

Caffe2\* PyTorch\* CNTK\* PaddlePaddle\*

### foundation

**Developers** 



#### **Pvthon** DAAL

Intel distribution optimized for machine learnina

All products, computer systems, dates, and figures are preliminary based on current expectations, and are subject to change without notice.

### Intel® Data Analytics

Acceleration Library (incl machine learnina)

**ANALYTICS, MACHINE & DEEP LEARNING PRIMITIVES** 

#### MKL-DNN

Open-source deep neural network functions for CPU / integrated graphics

#### **DEEP LEARNING GRAPH COMPILER**

#### Intel® nGraph™ Compiler (Alpha)

Open-sourced compiler for deep learning model computations optimized for multiple devices (CPU, GPU, NNP) from multiple frameworks (TF, MXNet, ONNX)

Formerly the Intel® Computer Vision SDK

<sup>\*</sup>Other names and brands may be claimed as the property of others. Developer personas show above represent the primary user base for each row, but are not mutually-exclusive

## HARDWARE Multi-purpose to purpose-built Al compute from cloud to device



### **Mainstream**



## —Training

Deep Learning













Most Al



## Ai compute continuum



device

### Cloud / Data Center



Large scale data centers such as public cloud or comms service providers, gov't and academia, large enterprise IT

### Edge



Small scale data centers, small business IT infrastructure, to few on-premise server racks and workstations

User-touch endpoint devices with lower power requirements such as laptops, tablets, smart home devices, drones

All products, computer systems, dates, and figures are preliminary based on current expectations, and are subject to change without notice.

## Deep learning inference accelerators





## Intel® FPGA

Custom deep learning inference





### Intel® Movidius™ VPU

Low power computer vision &





### Intel® Mobileye EyeQ

Autonomous driving inference





### Intel® GNA IP<sup>1</sup>

Ultra low power speech & audio inference









## Integrated graphics

Built-in deep learning inference





#### **Data Center**

Edge

Small scale clusters to a few on-premise server & workstations

Device

User-touch end-devices typically with lower power requirements

 ${}^{1}GNA = \underline{G}aussian \; Mixture \; Model \; and \; \underline{N}eural \; \underline{N}etwork \; \underline{A}ccelerator$ 

ll products, computer systems, dates, and figures are preliminary based on current expectations, and are subject to change without notice. Images are examples of intended applications but not an exhaustive list

## intel Solutions

Solution **Architects** 



#### **Al Solutions Catalog** (Public & Internal)









ARTIFICIAL INTELLIGENCE











### **TOOLKITS**

**Developers** 

libraries

Scientists





Open Visual Inference & Neural Network

#### **DEEP LEARNING DEPLOYMENT**

Optimized inference deployment on Intel VPUs for TensorFlow\* & Caffe\*

#### **REASONING**

Intel® Saffron™ AI

Cognitive solutions on CPU for anti-money laundering, predictive maintenance, more

**DEEP LEARNING FRAMEWORKS** 

#### DEEP LEAR Intel® Deep **Learning Studio**

Open-source tool to compress deep learning development cycle

#### Intel® Movidius™ SDK

Optimization toolkit for inference deployment on CPU/GPU/FPGA for TF, Caffe\* & MXNet\*

#### **MACHINE LEARNING LIBRARIES**

- **Pvthon**
- Scikit-learn
- Pandas
- NumPy
- Cart Random **Forest**

R

• e1071

#### Distributed

MlLib (on Spark)

**ANALYTICS, MACHINE & DEEP LEARNING PRIMITIVES** 

- Mahout

#### TensorFlow\* MXNet\*



mxnet



**cIDNN** 

Now optimized for CPU



Caffe

Caffe PYTORCH Caffe2\*

Optimizations in progress



CNTK\* PaddlePaddle\*



### **Id**oundation

Library **Developers** 



IT System **Architects** 



#### DAAL

Intel distribution optimized for machine learning

**Python** 

Intel® Data Analytics Acceleration Library (incl. machine learning)

#### **MKL-DNN**

Open-source deep neural network functions for CPU / integrated graphics

#### **DEEP LEARNING GRAPH COMPILER**

PvTorch\*

#### Intel® nGraph™ Compiler (Alpha)

Open-sourced compiler for deep learning model computations optimized for multiple devices (CPU, GPU, NNP) from multiple frameworks (TF, MXNet, ONNX)

#### AI FOUNDATION



Edge Device









**DEEP LEARNING ACCELERATORS** 







Inference





## **Future**

Driving AI forward through R&D, investment and policy leadership





✓ Image/Video/Audio

- ✓ Natural Language
- ✓ Autonomous Driving
- ✓ Reinforcement Learning
- ✓ Adversarial Learning
- ✓ Limited Precision/Sparsity
- ✓ More...

Over \$1B invested in Al innovators<sup>1</sup>

Partnering with #AI4GOOD to enrich the lives of every person on Earth through AI









## Intel® Xeon® processor Platform

#### **INFERENCE THROUGHPUT**





241x<sup>1</sup>

Intel® Xeon® Platinum 8180 Processor
higher Intel optimized Caffe GoogleNet v1 with Intel® MKL
inference throughput compared to
Intel® Xeon® Processor E5-2699 v3 with BVLC-Caffe

Inference and training throughput uses FP32 instructions

Intel® Xeon® Platinum 8180 Processor
higher Intel Optimized Caffe AlexNet with Intel® MKL training
throughput compared to
Intel® Xeon® Processor E5-2699 v3 with BVLC-Caffe

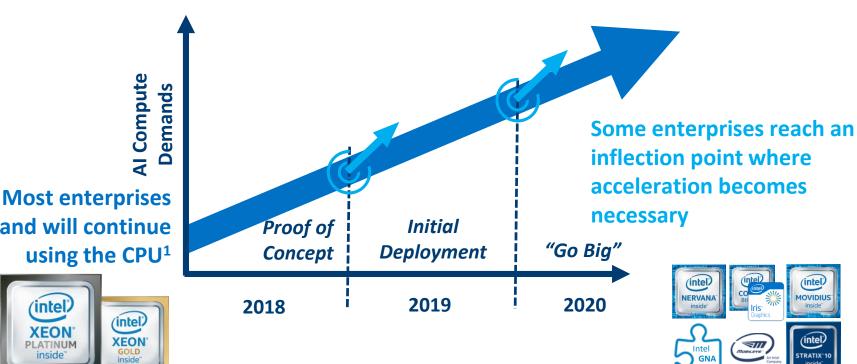


Deliver significant AI performance with hardware and software optimizations on Intel® Xeon® Scalable Family

¹ The benchmark results may need to be revised as additional testing is conducted. The results depend on the specific platform configurations and workloads utilized in the testing, and may not be applicable to any particular user's components, computer system or workloads. The results are not necessarily representative of other benchmarks and other benchmark results may show greater or lesser impact from mitigations. Software and workloads used in performance tests may have been optimized for performance tests, such as SYSmark and functions. Any change to any of those factors may cause the results to vary. You should consult office information and performance tests to assist you in fully evaluating your contemplated purchases, including the performance vary. You should consult office information and performance tests to assist you in fully evaluating your contemplated purchases. Including the performance vary. You should consult office information and performance vary. You should consult office information vary. You should consult office information and performance vary. You should consult office information vary. You should consult office information variety. You should consult office information variety in the variety of the performance variety

## Build on your AI Foundation

**Evaluate AI Acceleration Needs Moving Forward** 



are and will continue using the CPU<sup>1</sup>

AI Compute



## Intel® Xeon® Processor Scalable Family

Now build the AI you want on the CPU you know



your FOUNDATION for Al



### **Get maximum utilization**

running data center and AI workloads side-by-side



### **Break memory barriers**

in order to apply AI to large data sets and models



### **Train complex models**

through efficient scaling to many nodes



### **Access optimized tools**

including continuous performance gains for TensorFlow, MXNet, more



### Run in the cloud

including AWS, Microsoft, Alibaba, TenCent, Google, Baidu, more



### **Build on The premier AI portfolio**

from multi-purpose to purpose-built and cloud to device

## Intel® Nervana™ neural network

Fastest time-to-train for intensive deep learning environments Drocessor (NDP)



urpose-bult for deep learning



## **Blazingly-fast Data**

### Access

Using high bandwidth memory and separate compute and data pipelines



## High Degree of Parallelism

New numerical format (Bfloat16) for enhanced performance and conversion

## New Levels of Scalability

Massive hi-directional data transfer through

The Intel® Nervana™ Neural Network Processor is a future product that is not broadly available today

All products, computer systems, dates, and figures are preliminary based on current expectations, and are subject to change without notice.

## Deep learning frameworks

Popular DL Frameworks are now optimized for CPU



See installation guides at ai.intel.com/framework-optimizations/





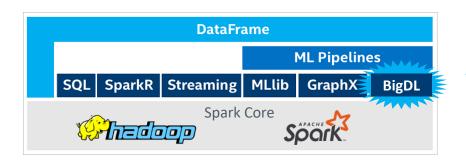




and more...

## **BigDL**

High Performance Deep Learning for FREE on CPU Infrastructure1



BigDL is a distributed deep learning library for Apache Spark\* that can run directly on top of existing Spark or Apache Hadoop\* clusters with direct access to stored data and tool/workflow consistency!



No need to deploy costly accelerators, duplicate data, or suffer through scaling headaches!



Feature Parity
with Caffe\* and
Torch\*



Lower TCO and improved ease of use with existing infrastructure



Deep Learning on Big
Data Platform,
Enabling Efficient
Scale-Out

software.intel.com/bigdl

pen-source software is available for download at no cost; 'free' is also contingent upon running on existing idle CPU infrastructure where the operating cost is treated as a 'sunk' cos

## Intel® Al academy

For developers, students, instructors and startups

Get smarter using online tutorials, webinars, student kits and support forums

Educate others using available course materials, hands-on labs, and more



Get 4-weeks FREE access to the Intel® AI DevCloud or use your existing Intel® Xeon® Processorbased cluster

Showcase your innovation at industry & academic events and online via the Intel AI community forum

software.intel.com/ai

# Learn more at ai.intel.com



## Fast & Efficient DL scaling on CPU

Intel® - SURFsara\* Research Collaboration - Multi-Node Intel® Caffe ResNet-50 Scaling Efficiency on 2S Intel® Xeon® Platinum 8160 Processing



90% scaling efficiency with up to 74% Top-1 accuracy on 256 nodes

Performance estimates were obtained prior to implementation of recent software patches and firmware updates intended to address exploits referred to as "Spectre" and "Meltdown." Implementation of these updates may make these results inapplicable to your device or system. Software and workloads used in performance tests may have been optimized for performance only on Intel microprocessors. Performance tests, such as SYSmark and MobileMark, are measured using specific computer systems, components, software, operations and functions. Any change to any of those factors may cause the results to vary. You should consult other informance tests to assist to assist to any of the performance of that product when combined with other products. For more complete information visit: <a href="http://www.intel.com/performance">http://www.intel.com/performance</a> Source: Intel measured as of June 2017

# Intel® Stratix® 10 FPGA

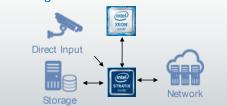
# inferenc **Efficient and low**



Scalable acceleration for deep learning inference in real-time with higher efficiency, and wide range of workloads & configurations

## flexibility

Reconfigurable for a variety of configurations & fast workload switching



### **Future-ready**

Future proof for new neural network topologies, arbitrary precision data types (FloatP32 => FixedP2, sparsity, weight sharing), inline & offload processing

Note: available as discrete or Xeon with Integrated FPGA (Broadwell Proof of Concept ) Configuration details on final slides

Up to **80%** reduction in power consumption (vs. Intel® Xeon® processor)<sup>1</sup>

Deterministic low latency, for real-time inline

processing of streaming data without

buffering (as low as <1ms<sup>2</sup> latency)

Projected latency (Googlete, PP11) batch size=2, memory banks=4, 0.3 ms)

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\*Projected latency (Googlete, PP11) batch size=2, 0.3 ms)

\*Projected latency (Googlete, PP11) batch size=2, 0.3 visit: http://www.intel.com/performance Source: Intel measured as of November 2016

voice interviews to the control of t Notice Revision #20110804

# Intel® Movidius™ Vision processing

Power-Efficient Image Processing, Computer Vision & Deep Learning for Devices

UNIT VPU DRONES

intel

MOVIDIUS



- Navigation •
- 3D Vol. Mapping •

SURVEITLE AND Censing .





- **Detection and Classification** 
  - Identification •
  - Multi-Nodal Systems •
  - Multi-Modal Sensing •

### **VIWEARABLES**

- **Detection, Tracking**
  - Recognition •
- Video, Image, Session Capture •



- Sense & Avoid
- GPS Denied Hovering
- Pixel Labeling
- Video, Image Capture



### **AR-VR HMD**

- **6DOF Pose, Position, mapping**
- **Gaze, Eye Tracking**
- **Gesture Tracking, Recognition**
- **See-Through Camera**

### **SMART HOME**



- Detection, Tracking
- Perimeter, Presence Monitoring
- Recognition, Classification
- Multi-Nodal Systems
- Multi-Modal Sensing
- Video, Image Capture



## Intel® Gaussian neural accelerator

Streaming Co-Processor for Low-Power Audio



## Ample throughput

For speech, language & other sensing inference

### Low power

<100 mW power consumption for always-on applications

## **Flexibility**

Gaussian Mixture Model (GMM) & Neural Network Inference support















Learn more: <a href="https://sigport.org/sites/default/files/docs/PosterFinal.pdf">https://sigport.org/sites/default/files/docs/PosterFinal.pdf</a>

# Intel integrated processor graphics

Built-in Deep Learning Inference Acceleration

**Iris**<sup>™</sup> Graphics

### **Ubiquity/Scalability**

- Shipped in > 1billion Intel SOCs
- Broad choice of performance/power offering across Intel® Atom™, Intel® Core™ and Intel® Xeon™ processors

### Media Leadership

- Intel® Quick Synch Video fixed function media blocks to improve power and performance
- Intel® Media SDK API that provides access to hardware-accelerated codecs

## Hardware integrat





### **Powerful & Flexible**

**Architecture** 

Rich data type support for 32bitFP, 16bitFP, 32bitInteger, 16bitInteger with SIMD multiply-accumulate instructions

### **Memory Architecture**

Shared memory architecture on die between CPU and GPU to enable lower latency and power

Software supp (MacOS (CoreML and MPS¹) Windows O/S (WinML) GIDNN Toolkit (Win, Linux)



## Visit www.mobileye.com

## Intel distribution for python

Advancing Python\* Performance Closer to Native Speeds



software.intel.com/intel-distribution-for-python

For developers using the most popular and fastest growing programming language for AI

## Easy, Out-of-the-box Access to High Performance Python

- Prebuilt, optimized for numerical computing, data analytics, HPC
- Drop in replacement for your existing Python (no code changes required)

# Drive Performance with Multiple Optimization Techniques

- Accelerated NumPy/SciPy/Scikit-Learn with Intel® MKL
- Data analytics with pyDAAL, enhanced thread scheduling with TBB, Jupyter\* Notebook interface, Numba, Cython
- Scale easily with optimized MPI4Py and Jupyter notebooks

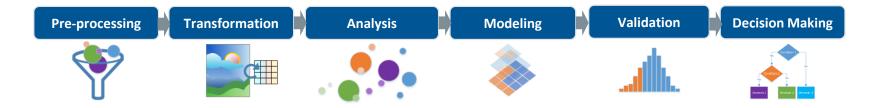
# Faster Access to Latest Optimizations for Intel Architecture

- Distribution and individual optimized packages available through conda and Anaconda Cloud
- Optimizations upstreamed back to main Python trunk

# Intel® Data Analytics Acceleration

High Performance Machine Learning and Data Analytics Library Library (Intel® DAAL)

Building blocks for all data analytics stages, including data preparation, data mining & machine learning



Open Source • Apache 2.0 License

Common Python, Java and C++ APIs across all Intel hardware

Optimized for large data sets including streaming and distributed processing

Flexible interfaces to leading big data platforms including Spark and range of data formats (CSV, SQL, etc.)

## Intel® MKL-dnN

Intel's Open-Source Math Kernel Library for Deep Neural Networks

For developers of deep learning frameworks featuring optimized performance on Intel hardware

#### Distribution Details

- Open Source
- Apache 2.0 License
- Common DNN APIs across all Intel hardware.
- Rapid release cycles, iterated with the DL community, to best support industry framework integration.
- Highly vectorized & threaded for maximal performance, based on the popular Intel® MKL library.

github.com/01org/mkl-dnn

#### **Examples:**

Direct 2D Convolution Local response normalization (LRN)

Rectified linear unit neuron activation (ReLU)

Maximum pooling

Inner product

## Intel® cldnn

<u>Compute Library for Deep Neural Networks on Intel Integrated Graphics</u>

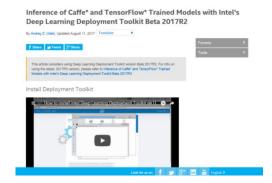
#### clDNN – Intel GPU DL acceleration middleware

- Open-sourced as of beginning May @ GitHub
- Official public page: <a href="https://01.org/cldnn">https://01.org/cldnn</a>
- Intel clDNN and Deep Learning Toolkit whitepaper
- DL Toolkit how to guide

### Frameworks integration

- Part of <u>Deep-Learning Deployment Toolkit</u>
- Part of Intel<sup>®</sup> OpenVINO™ Toolkit

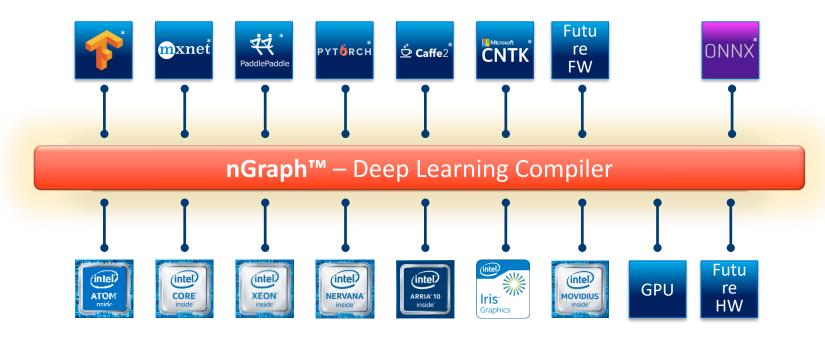




# Intel<sup>®</sup> ngraph™ compiler



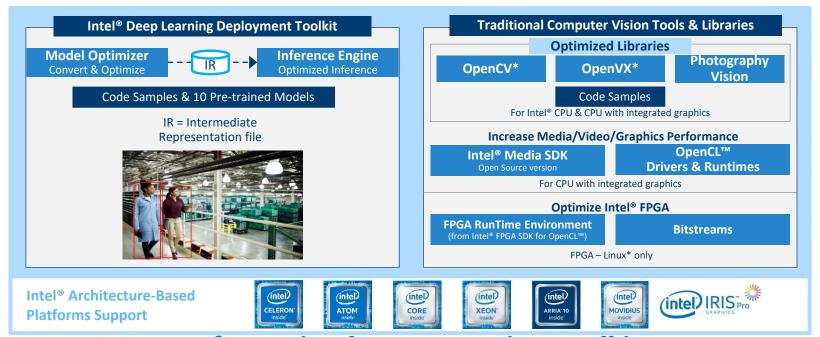
Open-source compiler enabling flexibility to run models across a variety of frameworks and hardware



<sup>\*</sup>Other names and brands may be claimed as the property of others. All products, computer systems, dates, and figures are preliminary based on current expectations, and are subje

## Openvino™ toolkit

Cross-Platform Tool to Accelerate Computer Vision & Deep Learning Inference Performance



software.intel.com/openvino-toolkit



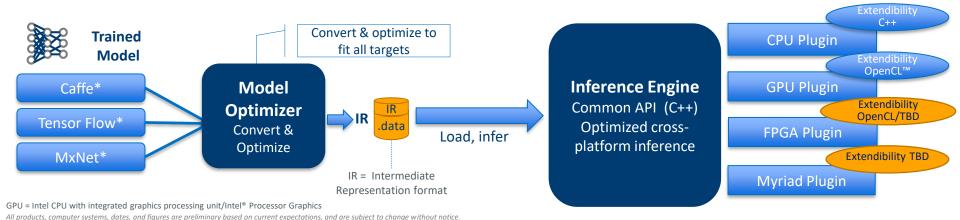
# Intel® Deep Learning Deployment

Take Full Advantage of the Power of Intel® Architecture for Deep Learning mizer

OOKIT Intel® Architecture for Deep Learning

### **Model Optimizer**

- What it is: Preparation step -> imports trained models
- Why important: Optimizes for performance/space with conservative topology transformations; biggest boost is from conversion to data types matching hardware.
- What it is: High-level inference API
- Why important: Interface is implemented as dynamically loaded plugins for each hardware type. Delivers best performance for each type without requiring users to implement and maintain multiple code pathways.

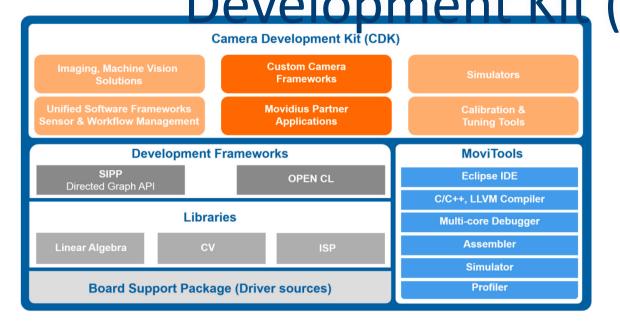


(intel) Al

## Intel<sup>®</sup> Movidius<sup>™</sup> Software

Comprehensive Software Development Suite for Video Processing Units (VPUs)

Development Kit (SDK)



## rapid prototyping

Remove time and complexity using built-in directed graph framework

## rapid prototyping

Remove time and complexity using built-in directed graph framework

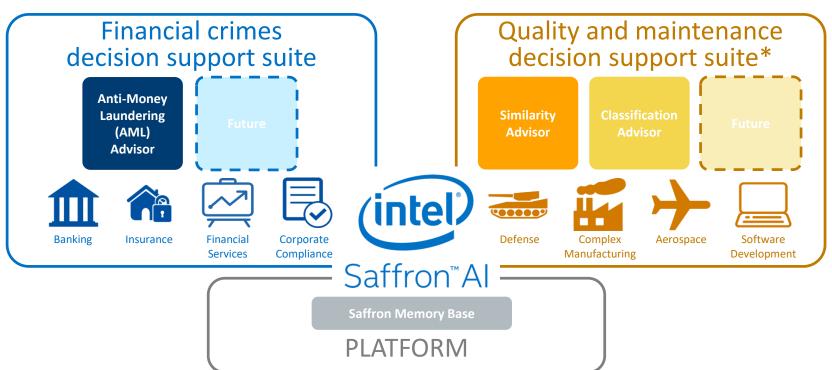
# Flexible development

From C/C++ to graphical development with rich tool suite

Learn more: https://uploads.movidius.com/1463156704-2016-04-29 MDK ProductBrief.pdf

## Intel<sup>®</sup> saffron<sup>™</sup> ai

Cognitive Reasoning Solutions for Complex Real-World Challenges

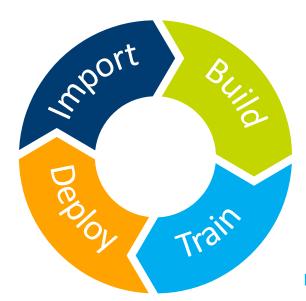


## Intel® Deep learning studio

Compress the Development Cycle to Accelerate Time-to-Solution

Data curation/processing
Data partitioning
Data labeling

Batch inference Model compression Inference deployment Export to edge devices



Multi-user collaboration Interactive sessions Model library

Fast training
Batch training
Experiment tracking
Multi-node distribution
Analytics & visualization
Hyperparameter optimization

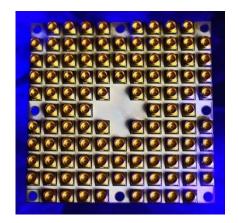
**Coming to the Intel® Deep Learning System** 

## Leading Al research

Choose a partner on the cutting-edge of AI breakthroughs



Neuromorphic Computing
Test Chip
Codenamed "Loihi"



Quantum Computing 49-Qubit Test Chip Codenamed "Tangle-Lake"



## Configuration details

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# Configuration details (Cont'd)

#### **Configuration: AI Performance – Software + Hardware**

INFERENCE using FP32 Batch Size Caffe GoogleNet v1 128 AlexNet 256.

The benchmark results may need to be revised as additional testing is conducted. The results depend on the specific platform configurations and workloads utilized in the testing, and may not be applicable to any particular user's components, computer system or workloads. The results are not necessarily representative of other benchmark results may show greater or lesser impact from mitigations. Performance tests, such as SYSmark and MobileMark, are measured using specific computer systems, components, software, operations and functions. Any change to any of those factors may cause the results to vary. You should consult other information and performance tests to assist you in fully evaluating your contemplated purchases, including the performance of that product when combined with other products. For more complete information visit http://www.intel.com/performance Source: Intel measured as of June 2017 Optimization Notice: Intel's compilers may or may not optimize to the same degree for non-Intel microprocessors for optimizations that are not unique to Intel microprocessors. These optimizations include SSE2, SSE3, and SSSE3 instruction sets and other optimizations. Intel does not guarantee the availability, functionality, or effectiveness of any optimization on microprocessors not manufactured by Intel. Microprocessor-dependent optimizations in this product are intended for use with Intel microprocessors. Certain optimizations not specific to Intel microarchitecture are reserved for Intel microprocessors. Please refer to the applicable product User and Reference Guides for more information regarding the specific instruction sets covered by this notice.

#### Configurations for Inference throughput

Platform :2 socket Intel(R) Xeon(R) Platinum 8180 CPU @ 2.50GHz / 28 cores HT ON , Turbo ON Total Memory 376.28GB (12slots / 32 GB / 2666 MHz),4 instances of the framework, CentOS Linux-7.3.1611-Core , SSD sda RS3WC080 HDD 744.1GB,sdb RS3WC080 HDD 1.5TB,sdc RS3WC080 HDD 5.5TB , Deep Learning Framework caffe version: a3d5b022fe026e9092fc7abc7654b1162ab9940d Topology:GoogleNet v1 BIOS:SE5C620.86B.00.01.0004.071220170215 MKLDNN: version: 464c268e544bae26f9b85a2acb9122c766a4c396 NoDataLayer. Measured: 1449.9 imgs/sec vs Platform: 2S Intel® Xeon® CPU E5-2699 v3 @ 2.30GHz (18 cores), HT enabled, turbo disabled, scaling governor set to "performance" via intel\_pstate driver, 64GB DDR4-2133 ECC RAM. BIOS: SE5C610.86B.01.01.0024.021320181901, CentOS Linux-7.5.1804(Core) kernel 3.10.0-862.3.2.el7.x86\_64, SSD sdb INTEL SSDSC2BW24 SSD 223.6GB. Framework BVLC-Caffe: <a href="https://github.com/BVLC/caffe">https://github.com/BVLC/caffe</a>, Inference & Training measured with "caffe time" command. For "ConvNet" topologies, dummy dataset was used. For other topologies, data was stored on local storage and cached in memory before training. BVLC Caffe (<a href="https://github.com/BVLC/caffe">https://github.com/BVLC/caffe</a>, prevision 2a1c552b66f026c7508d390b526f2495ed3be594

#### Configuration for training throughput:

Platform :2 socket Intel(R) Xeon(R) Platinum 8180 CPU @ 2.50GHz / 28 cores HT ON , Turbo ON Total Memory 376.28GB (12slots / 32 GB / 2666 MHz),4 instances of the framework, CentOS Linux-7.3.1611-Core , SSD sda RS3WC080 HDD 744.1GB,sdb RS3WC080 HDD 1.5TB,sdc RS3WC080 HDD 5.5TB , Deep Learning Framework caffe version: a3d5b022fe026e9092fc7abc765b1162ab9940d Topology:alexnet BIOS:SE5C620.86B.00.01.0004.071220170215 MKLDNN: version: 464c268e544bae26f9b85a2acb9122c766a4c396 NoDataLayer. Measured: 1257 imgs/sec vs Platform: 2S Intel® Xeon® CPU E5-2699 v3 @ 2.30GHz (18 cores), HT enabled, turbo disabled, scaling governor set to "performance" via intel\_pstate driver, 64GB DDR4-2133 ECC RAM. BIOS: SE5C610.86B.01.01.0024.021320181901, CentOS Linux-7.5.1804(Core) kernel 3.10.0-862.3.2.el7.x86\_64, SSD sdb INTEL SSDSC2BW24 SSD 223.6GB. Framework BVLC-Caffe: <a href="https://github.com/BVLC/caffe">https://github.com/BVLC/caffe</a>, Inference & Training measured with "caffe time" command. For "ConvNet" topologies, dummy dataset was used. For other topologies, data was stored on local storage and cached in memory before training. BVLC Caffe (<a href="https://github.com/BVLC/caffe">https://github.com/BVLC/caffe</a>, prevision 2a1c552b66f026c7508d390b526f2495ed3be594

# Configuration details (cont'd)

#### Intel® and SURFsara\* Research Collaboration MareNostrum4/BSC\* Configuration Details

\*MareNostrum4/Barcelona Supercomputing Center: https://www.bsc.es/

Compute Nodes: 2 sockets Intel® Xeon® Platinum 8160 CPU with 24 cores each @ 2.10GHz for a total of 48 cores per node, 2 Threads 96 GB of DDR4, Intel® Omni-Path Host Fabric Interface, dual-rail. Software: Intel® MPI Library 2017 Update 4Intel® MPI Library 2017 cache 1024K; L3 cache 33792K, Multi-EP. 10 Gbit Ethernet. 200 GB local SSD. Red Hat\* Enterprise Linux 6.7.

Intel® Caffe: Intel® version of Caffe; http://github.com/intel/caffe/, revision 8012927bf2bf70231cbc7ff55d Intel® MKL version: mklml Inx 2018.0.20170425; Intel® MLSL version: I mlsl 2017.1.016

Model: Topology specs from https://github.com/intel/caffe/tree/master/models/intel ize as stated in the performance chart

Time-To-Train: measured using "train" command. Data copied to memory or

#### Performance measured with:

export OMP NUM THREADS=44 (the remaining 4 cores are

...inreads per read All inreads per reads per read All inreads per reads per read All inreads per reads per read All inreads per read All inreads per read All inreads per read All inreads per reads per read All inreads per reads per read All inreads per reads p (ResNet-Sould troubling for wing communication) Matter I\_MPI managed the cluster before training. No input in the cluster before training. No input in the communication of the c The train and the state of the mpiexec.hydra -PSM2 -l -n \$SLURM JOB NUM NODES -ppn 1 -f hosts2 -genv OMP NUM THREADS 44 -enw UBSEL 1t -genv | MPI\_FABRICS tmi -genv | MPI\_HYDRA\_BRANCH\_COUNT .mdb > /dev/null ; cat /ilsvrc12\_val\_Imdb\_striped\_64/data.mdb > /dev/null ; SSLURM JOB NUM NODES -geny L ulimit -u 8192 : ulimit -a : num s/multinode/resnet 50 256 nodes 8k batch/solver poly quick large.prototxt engine "MKL2017"

SURFsara blog: https://blog.surf.nl earchers: Valeriu Codreanu, Ph.D. (PI).; Damian Podareanu, MSc; SURFsara\* & Vikram Saletore, Ph.D. (co-PI): Intel Corp.

\*SURFsara B.V. is the Dutch nation

# Configuration details (cont'd)

#### Intel® Arria 10 – 1150 FPGA energy efficiency on Caffe/AlexNet up to 25 img/s/w with FP16 at 297MHz

Vanilla AlexNet Classification Implementation as specified by http://www.cs.toronto.edu/~fritz/absps/imagenet.pdf, Training Parameters taken from Caffe open-source Framework are 224x224x3 Input, 1000x1 Output, FP16 with Shared Block-Exponents, All compute layers (incl. Fully Connected) done on the FPGA except for Softmax, Arria 10-1150 FPGA, -1 Speed Grade on Altera PCle DevKit with x72 DDR4 @ 1333 MHz, Power measured through on-board power monitor (FPGA POWER ONLY), ACDS 16.1 Internal Builds + OpenCL SDK 16.1 Internal Builds, Compute machine is an HP Z620 Workstation, Xeon E5-1660 at 3.3 GHz with 32GB RAM. The Xeon is not used for compute.